

MATH 451/551

Chapter 6. Joint Distribution

6.3 Expected Values

$$\begin{aligned} E(X) \\ E\{q(X)\} \end{aligned}$$

GuanNan Wang
gwang01@wm.edu





Expected Values

Let X and Y be random variables with joint probability mass function $f(x, y)$ if the random variables are discrete or joint probability density function $f(x, y)$ if the random variables are continuous. The expected value of $g(X, Y)$ is

$$E \{g(X, Y)\} = \begin{cases} \sum_x \sum_y g(x, y) f(x, y) & X, Y \text{ discrete} \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f(x, y) dy dx & X, Y \text{ continuous} \end{cases}$$

- The expected value of a function of two random variables $g(X, Y)$ is simply the sum or integral of the product of the functions $g(x, y)$ and $f(x, y)$.

Example 1



Example 1

Find $E(X + Y)$ for the discrete random variables X and Y with joint pmf

$$f(x, y) = \begin{cases} 0.2 & x = 1, y = 1 \\ 0.1 & x = 1, y = 2 \\ 0.3 & x = 1, y = 3 \\ 0.1 & x = 2, y = 1 \\ 0.1 & x = 2, y = 2 \\ 0.2 & x = 2, y = 3 \end{cases}.$$

$$\begin{aligned} E(\underline{X+Y}) &= \sum_{\underline{A}} \sum_{\underline{B}} g(x, y) f(x, y) \\ g(x, y) &= (1+1)*0.2 + (1+2)*0.1 + (1+3)*0.3 + (2+1)*0.1 + \\ &\quad (2+2)*0.1 + (2+3)*0.2 \\ &= 0.4 + 0.3 + 1.2 + 0.3 + 0.4 + 1.0 = 3.6 \end{aligned}$$

Example 2

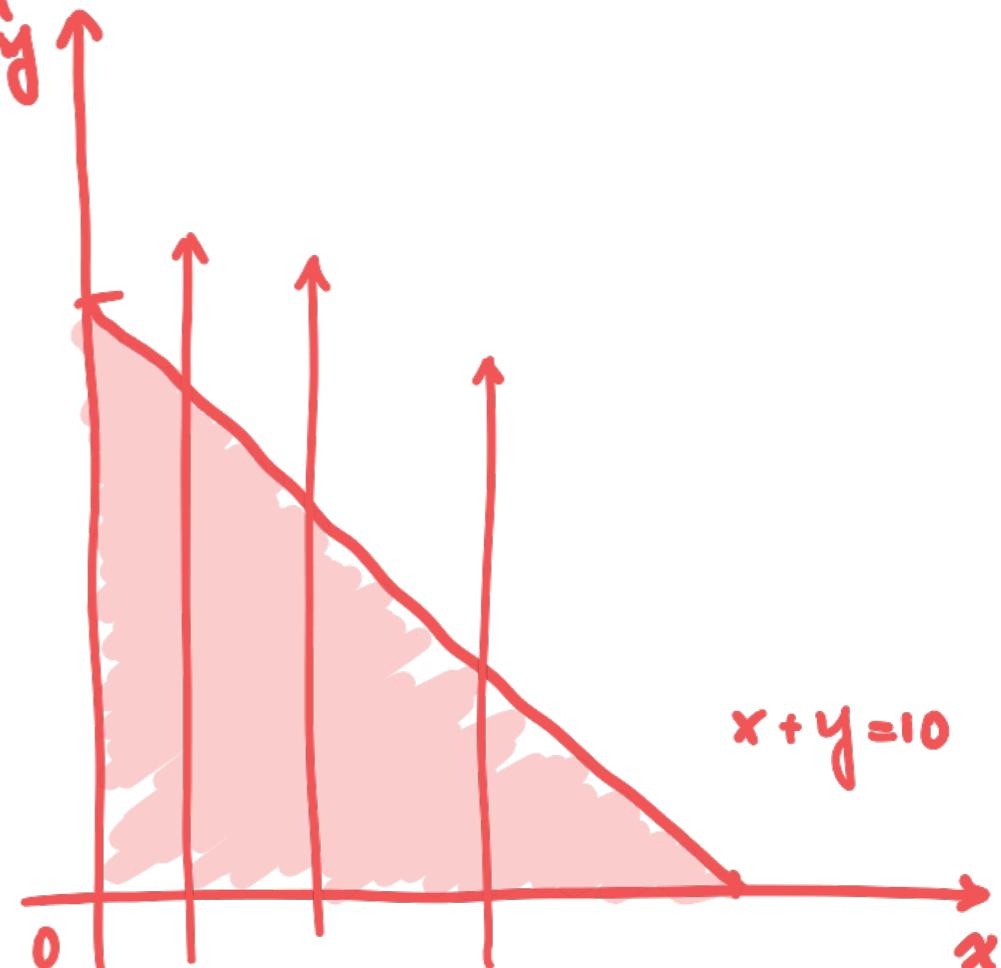


Example 2

Find $E(X^2 Y)$ for X and Y with joint pdf

$$f(x, y) = \frac{1}{50} \quad \underbrace{x > 0, y > 0, x + y < 10}_{\text{3}}$$

$$\begin{aligned} E(X^2 Y) &= \int_0^{10} \int_0^{10-x} g(x, y) f(x, y) dy dx \\ g(x, y) &= \int_0^{10} \int_0^{10-x} x^2 y \frac{1}{50} dy dx \\ &= \int_0^{10} \frac{x^2 (10-x)^2}{100} dx \\ &= \frac{100}{3} \end{aligned}$$



Theorem



Theorem 6.2

If X and Y are random variables, then

$$E\{g(X) + h(Y)\} = E\{g(X)\} + E\{h(Y)\}$$

for any functions g and h .

Proof: Let $f(x, y)$ denote the joint probability density fn of X & Y .

$$\begin{aligned} E\{g(x) + h(y)\} &= \iint_Q \{g(x) + h(y)\} f(x, y) dy dx \\ &= \iint_Q \{g(x)f(x, y) + h(y)f(x, y)\} dy dx \\ &= \underbrace{\iint_Q g(x)f(x, y) dy dx}_{\text{1st term}} + \underbrace{\iint_Q h(y)f(x, y) dy dx}_{\text{2nd term}} \\ &= E\{g(x)\} + E\{h(y)\} \end{aligned}$$

Theorem



Theorem 6.3

If X and Y are independent random variables, then

$$E\{g(X)h(Y)\} = E\{g(X)\}E\{h(Y)\}$$

for any functions g and h .

Proof. X & Y are independent $\Rightarrow f(x, y) = f_X(x)f_Y(y)$

$$\begin{aligned} E\{g(x)h(y)\} &= \iint_{\mathbb{R}^2} g(x)h(y)f(x, y) dy dx \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x)h(y) \underbrace{f_X(x)}_{\text{independent}} \underbrace{f_Y(y)}_{\text{independent}} dy dx \\ &= \underbrace{\int_{-\infty}^{\infty} g(x)f_X(x) dx}_{\text{independent}} * \underbrace{\int_{-\infty}^{\infty} h(y)f_Y(y) dy}_{\text{independent}} \\ &= E(g(X)) * E(h(Y)) \end{aligned}$$

Thank You



THANK YOU!

